

SECURITY CLASSIFIC OF THIS PAGE (When Data Entered) READ INSTRUCTIONS REPURT DOCUMENTATION PAGE BEFORE COMPLETING FORM 2. GOVT ACCESSION NO. 3. RECIPIENT'S CATALOG NUMBER REPORT NUMBER AFOSR TR- 78-0046 5. TYPE OF REPORT & PERIOD COVERED 4. TITLE (and Subtitle) MULTISTATE COHERENT SYSTEMS Interim FC FSU-Statistics-. AUTHOR(#) CONTRACT OR GRANT NUMBER(s) El-Neweihi, F. Proschan, J. Sethuraman 10. PROGRAM ELEMENT, PROJECT, TASK AREA & WORK UNIT NUMBERS 9. PERFORMING ORGANIZATION NAME AND ADDRESS Florida State University Department of Statistics V 61102F Tallahassee, FL 32306 1. CONTROLLING OFFICE NAME AND ADDRESS REPORT DAT Air Force Office of Scientific Research/NM Oct - 77 Bolling AFB, DC 20332 14. MONITORING AGENCY NAME & ADDRESS(If different from Controlling Office) SECURITY CLASS. (of this r UNCLASSIFIED 15a. DECLASSIFICATION DOWNGRADING SCHEDULE R-78-4046, 13888.12-M Approved for public release; distribution unlimited. FEB 15 1918 17. DISTRIBUTION STATEMENT (of the abstract entered in Block 20, if different from Report) 18. SUPPLEMENTARY NOTES 19. KEY WORDS (Continue on reverse side if necessary and identify by block number) multistate coherent systems, coherent systems, reliability, performance, IFRA, NBU, redundancy 20. ABSTRACT (Continue on reverse side if necessary and identify by block number) The vast majority of reliability analyses assume that components and system are in either of two states: functioning or failed. The present paper develops basic theory for the study of systems of components in which any of a finite number of states may occur, representing at one extreme perfect functioning and at the other extreme complete failure. We lay down axioms extending the standard notion of a coherent system to the new notion of a multistate coherent system. For such systems, we obtain deterministic and probabilistic properties for system performance which are analogous to well known results for coherent system reliability EDITION OF 1 NOV 65 IS OBSOLETE

400 277

UNCLASSIFIED
SECURITY CLASSIFICATION OF THIS PAGE (When Data Entered)

# AFOSR-TR- 78-0046

#### MULTISTATE COHERENT SYSTEMS

by

E. El-Neweihi<sup>1</sup>, F. Proschan<sup>2</sup>, and J. Sethuraman<sup>3</sup>

FSU Statistics Report M434 AFOSR Technical Report No. 76 USARO-D Technical Report No. 27

October, 1977

The Florida State University Department of Statistics Tallahassee, Florida 32306



Department of Statistics, University of Kentucky. Research sponsored by the Air Force Office of Scientific Research, AFSC, USAF, under Grant AFUSR-77-5522.

Research sponsored by Air Force Office of Scientific Research, AFSC, USAF, under Grant AFOSR-74-2581D.

<sup>3</sup>Research supported by the United States Army Research Office, Durham, under Grant No. DAA 29-76-G-0238.

Approved for public release;

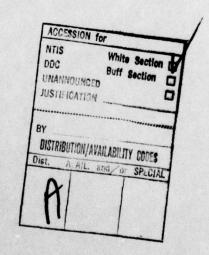
distribution unlimited. AIR FORCE OFFICE OF SCIENTIFIC RESEARCH (AFSC) MOTICE OF TRANSMITTAL TO DDC This technical report has been reviewed and isapproved for public release IAW AFR 190-12 (76). Distribution is unlimited. A. D. BLOSE Technical Information Officer

40

#### ABSTRACT

#### Multistate Coherent Systems.

The vast majority of reliability analyses assume that components and system are in either of two states: functioning or failed. The present paper develops basic theory for the study of systems of components in which any of a finite number of states may occur, representing at one extreme perfect functioning and at the other extreme complete failure. We lay down axioms extending the standard notion of a coherent system to the new notion of a multistate coherent system. For such systems we obtain deterministic and probabilistic properties for system performance which are analogous to well known results for coherent system reliability.



#### Multistate Coherent Systems.

#### 1. Introduction and Summary.

A central problem in reliability theory is to determine the relationship between the reliability of a complex system and the reliabilities of its components. Thus far, in practically all treatments of this problem, the system and its components are considered to be in either of two states: functioning (denoted by 1) and failed (denoted by 0). The theory of binary coherent structures has served as a unifying foundation for a mathematical and statistical theory of reliability for this dichotomous case. In fact, fairly complete solutions of various aspects of this problem have been obtained by Birnbaum, Esary, Saunders, Marshall, Barlow, and Proschan. See, for example, [1], [3], [4], [5], [6], and [7].

In many real life situations, however, the systems and their components are capable of assuming a whole range of levels of performance, varying from perfect functioning (denoted by level M, say) to complete failure (denoted by 0). In these situations, the dichotomous model is an oversimplification of the actual situation, and so models representing <u>multistate</u> systems and <u>multistate</u> components are much more useful in describing the performance of these systems in terms of the performance of their components.

Unfortunately, very little work has been done on this more general problem of multistate systems. Among the earlier papers treating aspects of multistate systems are [8], [9], [10], [11], [13], and [14]. With the exception of [10], and [11], these papers deal mainly with models for cannibalization, and barely touch on the performance of systems and components assuming more than two states. More recent and more sophisticated work on multistate systems has been performed by Barlow [2] and Ross [12]; however, their models are more specialized than ours, as we shall see.

The main purpose of the present paper is to develop an adequate general model and theory for the case in which both systems and components may assume any of an ordered set of states, say 0, 1, 2, ..., M; this theory generalizes coherent structure theory. We develop the concept of a multistate coherent structure as a generalization of the well known binary coherent structure. We then use this concept as a unifying foundation for the study of the relationship between the performance of a system and the performance of the components in the system. In forthcoming papers, we shall present treatments of various stochastic aspects of multistate systems with the aim of ultimately achieving a comprehensive theory analogous to coherent structure theory in the binary case (see [1]).

We now summarize the contents of this paper. Our formulation and treatment are similar to that of Barlow and Proschan [1] for the two state case. In Section 2 we present the notation and terminology used throughout the paper. In Section 3 we consider a system of n components. For each component and for the system itself, we can distinguish among say M + 1 states representing successive levels of performance ranging from perfect functioning (level M) down to complete failure (level 0). For component i,  $x_i$  denotes the corresponding state or performance level, i = 1, 2, ..., n; the vector  $\underline{x} = (x_1, ..., x_n)$  denotes the vector of states of components 1, ..., n. We assume that the state  $\phi$  of the system is a deterministic function of the states  $x_1, ..., x_n$  of the components. Thus  $\phi = \phi(\underline{x})$ , where  $\underline{x}$  takes values in  $S_n^n S = \{0, 1, ..., M\}$ , and  $\phi$  takes values in S. We define a multistate coherent structure as a natural generalization of the standard concept of a binary coherent structure by requiring three reasonable conditions that  $\phi$  must satisfy.

We then obtain deterministic relationships between the performance of a system and the performance of its components; these relationships are natural generalizations of well known results in the binary case. Thus we show that the

performance of a multistate coherent system is bounded below by the performance of a series system and bounded above by the performance of a parallel system.

We next present a decomposition identity useful in deriving inductive proofs and probabilistic properties for systems. Finally, we generalize the practical result that redundancy at the component level is better than redundancy at the system level.

In Section 4 we investigate the probabilistic aspects of multistate coherent systems. We relate in a probabilistic sense the performance of the system to the performance of its components, assumed statistically independent. Next the decomposition identity of Section 3 is used to obtain a corresponding decomposition identity for the performance function of the system. This decomposition identity is then used to show that system performance is a monotone increasing function of component performances. We end the section by obtaining bounds on system performance.

Finally, in Section 5 we study some dynamic aspects of multistate coherent systems. In earlier sections, we tacitly assume that time is fixed. In Section 5 we consider multistate coherent systems as operating over time. At time 0 the system and each of its components are in state M(corresponding to perfect functioning). As time passes, the performance level of components (and consequently of the system) deteriorates to lower levels until finally level 0 (complete failure) is reached. We define the concepts of IFRA and NBU stochastic processes introduced by Ross [12]. We present a different definition for an NBU stochastic process, and prove the analogue of the NBU closure theorem using a new characterization of the NBU property.

### 2. Notation and Terminology.

The vector  $\underline{\mathbf{x}} \equiv (\mathbf{x}_1, \dots, \mathbf{x}_n)$  denotes the vector of states of components 1, ..., n.

S = {0, 1, ..., M} denotes the set of possible states of both components and systems.

 $C = \{1, 2, ..., n\}$  denotes the set of component indices.

$$(j_i, \underline{x}) \equiv (x_1, ..., x_{i-1}, j, x_{i+1}, ..., x_n), \text{ where } j = 0, 1, ..., M.$$

$$(\cdot_i, \underline{x}) \in (x_1, \ldots, x_{i-1}, \cdot, x_{i+1}, \ldots, x_n).$$

 $\underline{y} < \underline{x}$  means  $y_i \le x_i$  for i = 1, ..., n, and  $y_i < x_i$  for some i.

 $x \vee y$  denotes max(x, y). and to notificall constraint of the political political

$$\underline{\mathbf{x}} \vee \underline{\mathbf{y}} \equiv (\mathbf{x}_1 \vee \mathbf{y}_1, \dots, \mathbf{x}_n \vee \mathbf{y}_n).$$

x A y denotes min(x, y). The sale and secondary transfers to not south

$$\underline{\mathbf{x}} \wedge \underline{\mathbf{y}} \equiv (\mathbf{x}_1 \wedge \mathbf{y}_1, \ldots, \mathbf{x}_n \wedge \mathbf{y}_n).$$

"increasing" is used in place of "nondecreasing", and "decreasing" is used in place of "nonincreasing".

When we say  $f(x_1, \ldots, x_n)$  is increasing we mean f is increasing in each argument.

Given a univariate distribution F, its complement 1-F is denoted by F.

quently of the system) deteriorates to lower levels until inmally level 0 (conplate failure) is reached. We define the concepts of IFRA end NGU sceenastic

tunctioning). As time passes, the performance level of components (and conse-

processes introduced by Ross [12]. We prosent a different definition for an

this atochastic process, and prove the analogue of the Niki closure theorem using

a new characteristing of the NAU proporty.

#### 3. Deterministic Properties of Multistate Coherent Systems.

Consider a system of n components. We assume that for both the system and its components we can distinguish among a finite number of distinct states representing various levels of performance, ranging from perfect functioning (state M) to complete failure (state 0). As time passes a component, starting in state M, deteriorates and enters state M-1, deteriorates further entering state M-2, etc., until ultimately it descends to state 0; a similar succession of decreasing state levels describes the system progression over time. (In a later paper, we consider the case of a continuous range of performance levels varying over the interval [0, 1].)

The performance  $x_i$  of component i assumes a value in the set  $S = \{0, 1, ..., M\}$ . We assume that the performance of the system depends deterministically on the performance of each of the components. Thus the state of the system is determined by a function  $\phi \colon S^n \to S$ . Given  $\underline{x}$ , the vector of component states, we may determine  $\phi(\underline{x})$ , the system state. The function  $\phi$  is called the <u>structure function</u> of the system.

The structure function  $\phi$  satisfies certain conditions that represent intuitively reasonable properties of systems encountered in practice. In the binary case the following two conditions are required for a system to be a coherent structure ([1], Def. 2.1, p. 6):

- (i). The function  $\phi(x)$  is increasing.
- (ii). Each component is <u>relevant</u> to the system; i.e., for each i there exists a vector  $(\cdot_i, \underline{x})$  such that  $\phi(l_i, \underline{x}) > \phi(l_i, \underline{x})$ . This means that the function  $\phi$  is not constant in the i<sup>th</sup> argument, i = 1, ..., n.

Condition (i) embodies the reasonable assumption that improving the performance of a component is not harmful to system performance. Condition (ii) eliminates from consideration components which have no effect on system performance. In our present multistate model, we stipulate three conditions:

Definition 3.1. A system of n components is said to be a <u>multistate coherent</u>

system (MCS) if its structure function  $\phi$  satisfies:

- (i)'. \$ is increasing.
- (ii)'. For level j of component i, there exists a vector  $(\cdot_i, \underline{x})$  such that  $\phi(j_i, \underline{x}) = j$  while  $\phi(\ell_i, \underline{x}) \neq j$  for  $\ell \neq j$ , i = 1, ..., n and j = 0, 1, ..., M.

  (iii)'.  $\phi(j) = j$  for j = 0, 1, ..., M.

Note that conditions (i)' and (ii)' generalize conditions (i) and (ii) in the binary case. Condition (iii)' is automatically satisfied in the binary case, but is not implied in the present multistate case by (i)' and (ii)'.

Some examples of MCS's are:

Example 3.1. A series system:  $\phi(\underline{x}) = \min_{1 \le i \le n} x_i$ .

Example 3.2. A parallel system:  $\phi(\underline{x}) = \max_{1 \le i \le n} x_i$ .

Example 3.3. A k-out-of-n structure:  $\phi(\underline{x}) = x_{(n-k+1)}$ , where  $x_{(1)} \le ... \le x_{(n)}$  is an increasing rearrangement of  $x_1, ..., x_n$ .

Example 3.4. Let  $P_1$ , ...,  $P_r$  be nonempty subsets of  $C = \{1, ..., n\}$  such that  $U P_i = C$  and  $P_i \Rightarrow P_j$ ,  $i \neq j$ . Let  $\phi(\underline{x}) = \max_{1 \leq j \leq r} \min_{i \in P_j} x_i$ . Then  $\phi(\underline{x})$  is a left  $i \in P_j$  MCS, and  $P_1$ , ...,  $P_r$  are called its min path sets.

Remark 3.1. The structures in Examples 3.1, 3.2, and 3.3 are natural generalizations of familiar basic structures in the binary case. They constitute special cases of the structure in Example 3.4, which in the binary case defines the most general binary coherent structure ([1], Chap. 1). The structure function of Example 3.4 is due to Barlow [2]. Since the structure functions of Example 3.4 satisfy conditions (i)', (ii)', and (iii)' of Definition 3.1, they constitute a

subclass of our MCS class. By examining some special cases it is easy to see that the class in Example 3.4 is actually a small subclass of our MCS. For instance, for a two component system, Example 3.4 yields only two systems: the parallel system and the series system. However for S = {0, 1, 2} there are more than 12 MCS's.

In the remainder of this section we investigate the structural properties of the MCS. We extend results obtained in the binary case ([1], Chap. 1) to the more general multistate case.

The following lemma gives a decomposition identity useful in carrying out inductive proofs. It holds for any multistate structure, not just for the MCS.

Lemma 3.1. The following identity holds for any n-component structure function  $\phi$ .

(3.1) 
$$\phi(\underline{x}) = \sum_{j=0}^{M} \phi(j_{i}, \underline{x}) I_{[x_{i}=j]} \text{ for } i = 1, ..., n,$$

where

$$I[x_i=j] = \begin{cases} I & \text{if } x_i=j \text{ odd , with respect to the last of the state of t$$

The proof is obvious and therefore omitted.

The following theorem gives simple bounds on MCS performance.

Theorem 3.1. Let  $\phi$  be the structure function of an MCS of n components.

Then

(3.2) 
$$\min_{1 \le i \le n} x_i \le \phi(\underline{x}) \le \max_{1 \le i \le n} x_i.$$

<u>Proof.</u> Let  $m = \max_{1 \le i \le n} x_i$ . Then  $\phi(\underline{x}) \le \phi(\underline{m})$  by the monotonicity of  $\phi$ . By condition (iii)' of Def. 3.1,  $\phi(\underline{m}) = m$ . The upper bound follows.

The proof establishing the lower bound is similar. |

Theorem 3.1 states that a parallel system yields the best performance of an MCS, and a series system yields the worst performance. Using this theorem, we will show similar probabilistic bounds in Section 4.

As in the binary case, we may define a dual structure for each multistate structure.

Definition 3.2. Let  $\phi$  be the structure function of a multistate system. The dual structure function  $\phi^D$  is given by:

(3.3) 
$$\phi^{D}(\underline{x}) = M - \phi(M - x_{1}, ..., M - x_{n}).$$

It is easy to verify that the dual of an MCS is an MCS.

Example 3.5. The dual of a series (parallel) system is a parallel (series) system. More generally, the dual of a k-out-of-n system is an (n - k + 1)-out-of-n system.

Design engineers have used the well known principle that redundancy at the component level is preferable to redundancy at the system level (all other things being equal). We present this principle in mathematical form along with a proof for MCS's.

Theorem 3.2. Let \$\phi\$ be the structure function of an MCS. Then

(i). 
$$\phi(\underline{x} \vee \underline{y}) \geq \phi(\underline{x}) \vee \phi(\underline{y})$$
.

(ii). 
$$\phi(\underline{x} \wedge \underline{y}) \leq \phi(\underline{x}) \wedge \phi(\underline{y})$$
.

Equality holds in (i) for all  $\underline{x}$  and  $\underline{y}$  if and only if the structure is parallel. Equality holds in (ii) for all  $\underline{x}$  and  $\underline{y}$  if and only if the structure is series.

<u>Proof.</u> (i).  $x_i \vee y_i \geq x_i$ , i = 1, ..., n. Thus  $\phi(\underline{x} \vee y) \geq \phi(\underline{x})$  since  $\phi$  is increasing. Similarly,  $\phi(\underline{x} \vee y) \geq \phi(\underline{y})$ . It follows that  $\phi(\underline{x} \vee y) \geq \max[\phi(\underline{x}), \phi(y)] \equiv \phi(\underline{x}) \vee \phi(\underline{y})$ .

(ii). A similar argument proves (ii).

If the structure is parallel (series), then equality in (i) ((ii)) is readily established.

Next assume  $\phi(\underline{x} \vee \underline{y}) = \phi(\underline{x}) \vee \phi(\underline{y})$  for all  $\underline{x}$  and  $\underline{y}$ . For each i, j, there exists  $(\cdot_i, \underline{x})$  such that  $\phi(j_i, \underline{x}) = j$  and  $\phi(x_i, \underline{x}) < j$  when  $x_i < j$ ,  $i = 1, \ldots, n$  and  $j = 0, 1, \ldots, M$  (by (ii)' of Def. 3.1). Since  $(j_i, \underline{x}) = (j_i, \underline{0}) \vee (0_i, \underline{x})$ , we have  $j = \phi(j_i, \underline{0}) = \phi(j_i, \underline{0}) \vee \phi(0_i, \underline{x})$ . It follows that  $\phi(j_i, \underline{0}) = j$  for  $i = 1, \ldots, n$  and  $j = 0, 1, \ldots, M$ . Now  $\phi(\underline{x}) = \phi(x_1, 0, \ldots, 0) \vee \phi(0, X_2, 0, \ldots, 0)$   $\vee \cdots \vee \phi(0, 0, \ldots, X_n) = x_1 \vee x_2 \vee \cdots \vee x_n \equiv \max_{1 \le i \le n} x_i$ . Therefore  $\phi$  is a parallel structure.

Finally, to prove necessity for equality in (ii), assume  $\phi(\underline{x} \wedge \underline{y}) = \phi(\underline{x}) \wedge \phi(\underline{y})$  for all  $\underline{x}$  and  $\underline{y}$ . Let  $\phi^D$  be the dual MCS of  $\phi$ . It is easy to show that  $\phi^D(\underline{x} \vee \underline{y}) = \phi^D(\underline{x}) \vee \phi^D(\underline{y})$  for all  $\underline{x}$  and  $\underline{y}$ . Hence by (i),  $\phi^D$  is a parallel structure. Therefore  $\phi$  is a series structure.

In binary coherent structure theory, the concepts of minimal path vectors and minimal cut vectors play a crucial role. The analogue in MCS theory is the concept of critical connection vectors. Using this concept we can represent the state of a MCS in terms of its critical connection vectors.

Definition 3.3. A vector  $\underline{\mathbf{x}}$  is said to be a connection vector to level  $\underline{\mathbf{j}}$  if  $\phi(\underline{\mathbf{x}}) = \underline{\mathbf{j}}$ ,  $\underline{\mathbf{j}} = 0$ , 1, ..., M.

Definition 3.4. A vector  $\underline{x}$  is said to be an upper critical connection vector to level  $\underline{j}$  if  $\phi(\underline{x}) = \underline{j}$  and  $\underline{y} < \underline{x}$  implies  $\phi(\underline{y}) < \underline{j}$ ,  $\underline{j} = 1, \ldots, M$ .

Similarly, we can define a lower connection vector to level j, j = 0, 1, ..., M-1.

The existence of such critical connection vectors is guaranteed by the conditions of Definition 3.1.

Let  $\underline{x}$  be an upper critical connection vector to level j. Define  $C_j(\underline{x}) = \{i: x_i \geq j\}$ . Obviously  $C_j(\underline{x})$  is a non-empty subset of  $C = \{1, \ldots, n\}$ . For  $j = 1, \ldots, M$ , let  $C_j = \{C_j(x): \underline{x} \text{ is an upper critical connection vector to level j}$ . Then the following lemma shows that  $C_j$  enjoys a property similar to that enjoyed by the minimal path sets and the minimal cut sets in the binary case.

Lemma 3.2. For j = 1, ..., M,

- (i). UC<sub>j</sub> = {1, 2, ..., n}.
- (ii). If A and B are two different members of  $C_i$ , then A  $\Rightarrow$  B.

The proof follows readily from Def. 2.1 and hence is omitted.

For j = 1, ..., M, let  $y_1^j$ , ...,  $y_{n_j}^j$  be the upper critical connection vectors to level j, where  $y_T^j = (y_{1T}^j, ..., y_{nT}^j)$ . The following theorem, stated without proof, enables us to determine the state of an MCS using its upper critical connection vectors.

Theorem 3.3. Let  $\phi$  be the structure function of an MCS. Let  $\underline{y}_1^j$ , ...,  $\underline{y}_{nj}^j$  be its upper critical connection vectors to level j, j = 1, ..., M. Then

$$\phi(\underline{x}) \geq j$$

if and only if  $\underline{x} \ge \underline{y}_{\ell}^{j}$  for some  $\ell$ ,  $1 \le \ell \le n_{j}$ .

Theorem 3.3 is used in Section 4 to establish bounds on the system performance function.

#### 4. Stochastic Performance of Multistate Coherent Systems.

In Section 3 we discussed deterministic aspects of MCS's. In this section, we determine the relationship between the stochastic performance of the system and the stochastic performances of its components. We also obtain bounds on system performance which are particularly useful when exact system performance is difficult to evaluate.

Let X denote the random state of component i, with

(4.1) 
$$P[X_{i} \leq j] = P_{i}(j),$$

j = 0, 1, ..., M and i = 1, ..., n. P<sub>i</sub> represents the performance distribution of component i. Clearly,

$$P_{i}(j) = \sum_{k=0}^{j} P_{ik}$$

$$P_{i}(M) = \sum_{k=0}^{M} p_{ik} = 1,$$

for i = 1, ..., in component independence of the component; ..., i rol

Let  $\underline{X} = (X_1, \ldots, X_n)$  be the random vector representing the states of components 1, ..., n, where the  $X_1, \ldots, X_n$  are assumed to be statistically mutually independent. Then  $\phi(\underline{X})$  is the random variable representing the system state of the MCS having structure function  $\phi$ , with

$$P[\phi(X) = j] = p_{j}, j = 0, 1, ..., M,$$

$$(4.2)$$

$$P[\phi(X) \le j] = P(j), j = 0, 1, ..., M.$$

P represents the <u>performance distribution of the system</u>. Let  $h = E \phi(X)$ ; we may express h as follows:

since h is a function of the  $P_1$ , ...,  $P_n$ . We may also express h alternatively:

$$h \equiv h_{\underline{p}}(p_1, p_2, ..., p_n),$$

where  $\underline{p}_i \equiv (p_{i0}, p_{i1}, \dots, p_{iM})$  for  $i = 1, \dots, n$ . In either case we call h the <u>performance function</u> of the system. We shall omit the subscript on h when no confusion will result.

The following identity expresses a system performance function of n components in terms of system performance functions of n-1 components.

Lemma 4.1. The following identity holds for h:

(4.3) 
$$h(\underline{p}_1, \underline{p}_2, ..., \underline{p}_n) = \sum_{j=0}^{M} p_{ij} h(j_i; \underline{p}_1, ..., \underline{p}_n), i = 1, ..., n,$$

where 
$$h(j_i; \underline{p}_1, \underline{p}_2, ..., \underline{p}_n) = E \phi(j_i, \underline{x}) = E \phi(x_1, ..., x_{i-1}, j, x_{j+1}, ..., x_n)$$
.

Proof. By Lemma 3.1 and the mutual independence of the components, we have:

$$E \phi(\underline{X}) = \sum_{j=0}^{M} E I_{[X_{i}=j]} \cdot E \phi(j_{i}, \underline{X}).$$

The following theorem shows that h is strictly increasing in each  $p_{ij}$  for j > 0.

Theorem 4.1. Let  $h(\underline{p}_1, \ldots, \underline{p}_n)$  be the performance function of an MCS. Let  $0 < p_{ij} < 1$  for  $i = 1, \ldots, n$  and  $j = 0, 1, \ldots, M$ . Then  $h(\underline{p}_1, \ldots, \underline{p}_n)$  is strictly increasing in  $p_{ij}$ ,  $i = 1, \ldots, n$  and  $j = 1, \ldots, M$ .

**Proof.** From (4.3) and the fact that  $\sum_{j=0}^{M} p_{ij} = 1$ , and i = 1, ..., n, we have

$$h(p_1, ..., p_n) = \sum_{j=1}^{M} p_{ij} E[\phi(j_i, \underline{x}) - \phi(0_i, \underline{x})].$$

Thus

$$\frac{\partial h}{\partial P_{ij}} = E[\phi(j_i, X) - \phi(0_i, X)], i = 1, ..., n \text{ and } j = 1, ..., M.$$

Since  $\phi$  is increasing,  $E[\phi(j_i, \underline{X}) - \phi(0_i, \underline{X})] \ge 0$ . In addition,  $\phi(j_i, \underline{x}^0) - \phi(0_i, \underline{x}^0) > 0$  for some  $(\cdot_i, \underline{x}^0)$  since the structure is an MCS. Since  $0 < p_{ij} < 1$  for all i and j, we have  $E[\phi(j_i, \underline{X}) - \phi(0_i, \underline{X})] > 0$ . Thus the desired result follows.  $| \cdot |$ 

Next we obtain properties of h as a function of the  $P_1$ , ...,  $P_n$ . First we show that  $h(P_1, \ldots, P_n)$  is monotone increasing with respect to stochastic ordering. A similar result is proved by Barlow [2] for his subclass of MCS's (see Ex. 3.4). Our proof of the more general result is simpler.

Theorem 4.2. Let  $P_i$ ,  $P_i'$  be two possible performance distributions for component i, i = 1, ..., n. Assume  $P_i(j) \ge P_i'(j)$  for j = 0, 1, ..., M and i = 1, ..., n. Let P(P') be the corresponding system performance distribution. Then

(ii). 
$$h(P_1, ..., P_n) \le h(P_1, ..., P_n)$$
.

<u>Proof.</u> Let  $X_1, \ldots, X_n(X_1', \ldots, X_n')$  be mutually independent random variables having distribution functions  $P_1, \ldots, P_n(P_1', \ldots, P_n')$  respectively. Then for  $i = 1, \ldots, n$ ,  $P_i(j) \ge P_i'(j)$  for  $j = 0, 1, \ldots, M$  implies that  $X_i < X_i'$ . Since  $\phi$  is increasing,  $\phi(\underline{X}) < \phi(\underline{X}')$ . The desired results in (i) and (ii) follow immediately.

Similarly, we relate properties of P, the system performance distribution, to properties of h, the system performance function, or to properties of the  $P_{ij}$ . As examples we state the following two straightforward results.

Lemma 4.2. Let h be the performance function and P be the performance distribution of an MCS. Then

$$h = \sum_{j=0}^{M-1} \overline{P}(j), \text{ where } \overline{P}(j) \equiv 1 - P(j).$$

<u>Proof.</u> Since  $\phi(\underline{X})$  is a nonnegative integer valued random variable, then  $\mathbb{E} \phi(\underline{X}) = \sum_{j=0}^{\infty} P[\phi(\underline{X}) > j]$ , yielding the desired result. | | |A decomposition identity is given in:

Theorem 4.3. Let \$\phi\$ be the structure function of a MCS. Then

(4.4) 
$$P[\phi(\underline{X}) \ge \ell] = \sum_{j=0}^{M} P_{ij} P[\phi(j_i, \underline{X}) \ge \ell], \ell = 1, ..., M.$$

Proof. By the law of total probability, we have:

$$P[\phi(\underline{X}) \geq \ell] = \sum_{j=0}^{M} P[\phi(\underline{X}) \geq \ell | X_{i} = j] P[X_{i} = j].$$

Since the components are mutually independent, (4.4) follows immediately. | |

Next we obtain bounds on both the system performance function and the

system performance distribution. Using Theorem 3.1, we establish:

Theorem 4.4. Let P be the performance distribution and h be the performance function of an MCS. Let  $P_i$  be the i<sup>th</sup> component performance distribution for i = 1, ..., n. Then for j = 0, 1, ..., M - 1:

(i). 
$$\prod_{i=1}^{n} P_{i}(j) \le P(j) \le 1 - \prod_{i=1}^{n} \overline{P}_{i}(j)$$
,

(ii). 
$$\sum_{j=1}^{M} \prod_{i=1}^{n} \overline{P}_{i}(j-1) \leq h \leq \sum_{j=1}^{M} [1 - \prod_{i=1}^{n} P_{i}(j-1)].$$

<u>Proof.</u> (i). By Theorem 3.1 we have  $\min_{1 \le i \le n} X_i \le \phi(X) \le \max_{1 \le i \le n} X_i$ . Since  $X_1, \ldots, X_n$  are mutually independent, (i) follows immediately.

(ii). We use the fact that  $h = \sum_{j=1}^{M} P[\phi(X) \ge j]$  and the bounds on  $P[\phi(X) \ge j]$ . ||

Next we illustrate how we can use the upper critical connection vectors to establish bounds on the system performance distribution P and consequently on system performance function h. Let  $y_1^j$ , ...,  $y_{nj}^j$  be the upper critical connections to level j, j = 0, 1, ..., M. Let  $A_r^j$  denote the event that  $\underline{X} \ge y_r^j$ ,  $\underline{Y}_r^j$ ,  $\underline{Y}_r^j$  and  $\underline{Y}_r^j$  by Theorem 3.3, it follows that  $P[\phi(\underline{X}) \ge j] = P[\underbrace{U}_r^j A_r^j]$ . Let  $S_k = \sum_{i=1}^{n} P[A_{i_1} \cap A_{i_2} \cap \cdots \cap A_{i_k}]$ . By the inclusion-exclusion principle,  $1 \le i_1 \le i_2 \le \cdots \le i_k \le n_j$ 

$$P[\phi(X) \ge j] = \sum_{k=1}^{n_j} (-1)^{k-1} S_k.$$

Thus

$$P[\phi(\underline{X}) \geq j] \leq S_1 = \sum_{r=1}^{n_j} P[A_r],$$

$$P[\phi(\underline{X}) \geq S_1 - S_2,$$

er salari Sjeka

and so on, constituting upper and lower bounds on  $P[\phi(\underline{X}) \geq j] = \overline{P}(j-1)$  for  $j = 1, \ldots, n$ . Since  $h = \sum_{j=1}^{M} P[\phi(\underline{X}) \geq j]$ , we automatically have upper and lower bounds on h also. Note that  $P(A_{\underline{r}}^j) = P[\underline{X} \geq \underline{y}_{\underline{r}}^j] = P[X_1 \geq y_{1\underline{r}}^j, X_2 \geq y_{2\underline{r}}^j, \dots, X_n \geq y_{n\underline{r}}^j] = \prod_{j=1}^{M} P[X_1 \geq y_{1\underline{r}}^j]$  for  $1 \leq r \leq n_j$  and  $j = 1, \ldots, M$ .

Following was [12], as present

as established and the survivation of

#### 5. Dynamic Models for Multistate Coherent Systems.

In the last two sections, we have studied deterministic and probabilistic properties of MCS's at a fixed point in time. Now we consider dynamic models, i.e., models in which the state of the system and of its components vary over time. At time 0, the system and each of its components are in state M (perfect functioning). As time passes, the performance of each component (and consequently of the system itself) deteriorates to successively lower levels, until ultimately failure occurs (level 0 is attained).

In the binary case, the length of time during which a component or system functions is called the <u>lifelength</u> of the component or system; these lifelengths are nonnegative random variables. The corresponding lifelength distributions have been classified according to various notions of aging. See, e.g., [1]. Two of the important classes of life distributions are the increasing failure rate average (IFRA) class and the new better than used (NBU) class. Closure of these classes under various basic reliability operations, such as convolution of distributions and formation of binary coherent systems, are investigated in [1]. In this section we investigate generalizations of these useful concepts in the multistate case.

Let  $\{X_i(t), t \ge 0\}$  denote the stochastic process representing component stat i at time t as t varies over the nonnegative real numbers, for i=1,...,n. The stochastic process  $\{\phi(\underline{X}(t)), t \ge 0\}$  represents the corresponding system state as t varies from 0 to  $\phi$ , where  $\underline{X}(t) = (X_1(t), \ldots, X_n(t))$ . We assume  $X_i(0) = M$ ,  $i = 1, \ldots, n$ . We assume as before that components are mutually statistically independent; thus the processes  $\{X_i(t), t \ge 0\}$ ,  $i = 1, \ldots, n$ , are also mutually independent.

Following Ross [12], we present:

Definition 5.1. The stochastic process  $\{X_i(t), t \ge 0\}$  is said to be an IFRA process if  $T_i^j \equiv \inf\{t: X_i(t) \le j\}$  is an IFRA random variable for j = 0,  $1, \ldots, M-1$ .

In a similar fashion we may define an IFRA process for the system:  $\{\phi(\underline{X}(t)), t \geq 0\}$ . Note that in the binary case,  $T_i^0$  is simply the lifelength of component i.

The following theorem is due to Ross [12].

Theorem 5.1. The Generalized IFRA Closure Theorem. Let  $\{X_i(t), t \ge 0\}$ , i = 1, ..., n, be independent IFRA processes and  $\phi$  an increasing structure function. Then  $\{\phi(\underline{X}(t)), t \ge 0\}$  is an IFRA process.

We now give a definition for NBU stochastic processes different from the one given by Ross [12]. We then derive a simple characterization for our NBU stochastic processes. Using this characterization, we give a simple proof of a generalized NBU closure theorem.

Definition 5.2. The stochastic process  $\{X_i(t), t \ge 0\}$  is an NBU stochastic process if  $T_i^j$  is an NBU random variable for j = 0, 1, ..., M - 1.

In a similar fashion we may define an NBU stochastic process  $\{\phi(\underline{X}(t)), t \ge 0\}$  for the system.

The following lemma gives a simple characterization for an NBU process

(as well as for an NBU random variable). We omit the simple proof.

Lemma 5.1. The stochastic process  $\{X(t), t \ge 0\}$  is NBU if and only if for all  $s \ge 0$  and  $t \ge 0$ :

$$X(s + t) \stackrel{\text{st}}{<} \min(X'(s), X'(t)),$$

where X'(s) and X'(t) are two independent random variables having the same distributions as X(s) and X(t) respectively.

We may now prove the main result of this section.

Theorem 5.2. Let  $\phi$  be the structure function of an MCS having n components. Let  $\{X_i(t), t \ge 0\}$ , i = 1, ..., n, be independent NBU stochastic processes. Then  $\{\phi(X(t)), t \ge 0\}$  is an NBU stochastic process.

<u>Proof.</u> For arbitrary fixed  $s \ge 0$  and  $t \ge 0$ , let  $X_1'(s), \ldots, X_n'(s), X_1'(t), \ldots, X_n'(t)$  be mutually independent random variables having the same distributions as  $X_1(s), \ldots, X_n(s), X_1(t), \ldots, X_n(t)$  respectively. Since  $\{X_i(t), t \ge 0\}$  is an NBU process, we have by Lemma 5.1:

$$X_{i}(s + t) < \min(X_{i}(s), X_{i}(t)), i = 1, ..., n.$$

Since \( \psi \) is increasing, it follows that

$$\phi(\underline{X}(s+t)) \stackrel{st}{<} \phi(\min(\underline{X}'(s), \underline{X}'(t)).$$

By Theorem 3.2,  $\phi(\min(\underline{X}'(s), \underline{X}'(t))) < \min(\phi(\underline{X}'(s)), \phi(\underline{X}'(t)))$ . Thus  $\phi(X(s+t)) < \min(\phi(X'(s)), \phi(X'(t)))$ . Using Lemma 5.1 again, the desired result follows.

Remark 5.1. The useful characterization of Lemma 5.1 makes our proof of the generalized NBU closure theorem in the binary case simpler than the proof given in [1].

#### REFERENCES

- [1] Barlow, R. E. and Proschan, F. (1975). Statistical Theory of Reliability and Life Testing. Holt, Rinehart and Winston, New York.
- [2] Barlow, R. E. (1977). Coherent systems with multi-state components. ORC 77-5, Operations Research Center, University of California, Berkeley.
- [3] Birnbaum, Z. W., Esary, J. D. and Saunders, S. C. (1961). Multicomponent systems and structures and their reliabilities. <u>Technometrics</u>, 3, 55-77.
- [4] Birnbaum, Z. W. and Esary, J. D. (1965). Some inequalities for reliability functions. Proc. of the Fifth Berkeley Symposium on Math. Statistics and Probability, 271-283.
- [5] Birnbaum, Z. W. (1960). On the probabilistic theory of complex structures.

  Proc. of the Fourth Berkeley Sumposium on Math. Statistics and

  Probability, 49-55.
- [6] Birnbaum, Z. W., Esary, J. D. and Marshall, A. W. (1966). Stochastic characterization of wearout for components and systems. <u>Ann. Math.</u> <u>Statist.</u> 37, 816-825.
- [7] Esary, J. D. and Proschan, F. (1963). Coherent structures of non-identical components. <u>Technometrics</u>, <u>5</u>, 191-209.
- [8] Hirsch, W. M. et al (1968). Cannibalization in multicomponent systems and the theory of reliability. Naval Research Logistics Quarterly, 15, 331-359.
- [9] Hochberg, M. (1973). Generalized multicomponent systems under cannibalization. Naval Research Logistics Quarterly, 20, 585-605.
- [10] Murchland, J. D. (1975). Fundamental concepts and relations for reliability analysis of multistate systems. Reliability and Fault Tree Analysis, ed. by R. E. Barlow, J. Fussell, and N. D. Singpurwalla.
- [11] Postelnicu, V. (1970). Nondichotomic multicomponent structures.

  Bulletin de la Societé des Sciences Mathématiques de la R. S.

  Roumaine, Tom. 14 (62), nv. 2, 209-217.
- [12] Ross, S. (1977). Multi-valued state component reliability systems.

  Technical Report, Department of Industrial Engineering and Operations Research, University of California, Berkeley.
- [13] Simon, R. M. (1970). Optimal cannibalization policies for multi-component systems. SIAM J. Appl. Math., 19, 700-711.
- [14] Simon, R. M. (1972). The reliability of multicomponent systems subject to cannibalization. <u>Naval Research Logistics Quarterly</u>, 19, 1-14.

# UNCLASSIFIED SECURITY CLASSIFICATION OF THIS PAGE

_	REPORT DOCUMENTATION PAGE		
1.	REPORT NUMBER 2. GOVT ACCESSION NO. FSU No. M434 AFOSR No. 76 USARO-D No. 27	3. RECIPIENT'S CATALOG NUMBER	
4.	TITLE (and subtitle)	5. TYPE OF REPORT & PERIOD COVERED	
	Multistate Coherent Systems'	Technical Report	
		6. PERFORMING ORG. REPORT NUMBER	
		FSU Statistics Report M434	
7.	AUTHOR(s)	8. CONTRACT OR GRANT NUMBER(s)	
	E. El-Neweihi, F. Proschan, J. Sethuraman	AFOSR-77-3322, AFOSR-74-2581D, DAA 29-76-G-0238	
9.	PERFORMING ORGANIZATION NAME & ADDRESS	10. PROGRAM ELEMENT, PROJECT, TASK AREA & WORK UNIT NUMBERS	
	The Florida State University Department of Statistics Tallahassee, Florida 32306		
1.	CONTROLLING OFFICE NAME & ADDRESS	12. REPORT DATE	
	The U.S. Air Force	October, 1977	
	Air Force Office of Scientific Research 1400 Wilson Boulevard	13. NUMBER OF PAGES	
	Arlington, Virginia 22209	19	
	United States Army Research Office-Durham P.O. Box 12211 Research Triangle Park, N.C. 27709		
1.	MONOTORING AGENCY NAME & ADDRESS (if different from Controlling Office)	15. SECURITY CLASS (of this report)	
	different from controlling office)	Unclassified	
		15a. DECLASSIFICATION/DOWNGRADING SCHEDULE	
	DISTRIBUTION STATEMENT (of this report)	IBUTION STATEMENT (of this report)	
	Approved for public release; distribution of		
7.	DISTRIBUTION STATEMENT (of the abstract ent	tered in Block 20, if different from report)	
8.	SUPPLEMENTARY NOTES		

#### 19. KEY WORDS

Multistate coherent systems, coherent systems, reliability, performance, IFRA, NBU, redundancy

## UNCLASSIFIED SECURITY CLASSIFICATION OF THIS PAGE

#### 20. ABSTRACT

The vast majority of reliability analyses assume that components and system are in either of two states: functioning or failed. The present paper develops basic theory for the study of systems of components in which any of a finite number of states may occur, representing at one extreme perfect functioning and at the other extreme complete failure. We lay down axioms extending the standard notion of a coherent system to the new notion of a multistate coherent system. For such systems we obtain deterministic and probabilistic properties for system performance which are analogous to well known results for coherent system reliability.